

Automation of Structure Determination by Macromolecular Crystallography

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Automation of structure determination

Automation...

makes straightforward cases accessible to a wider group of structural biologists

makes difficult cases more feasible for experts

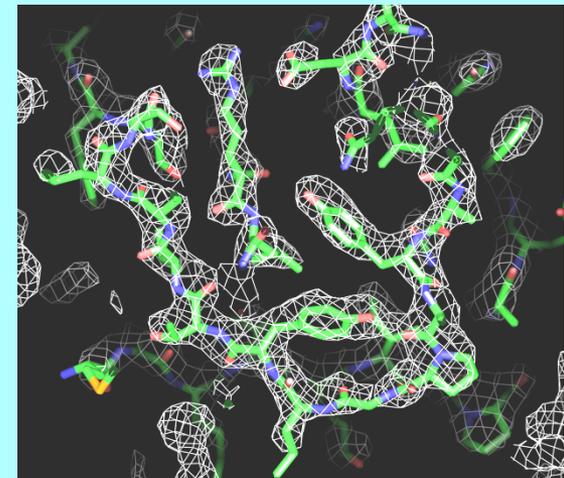
can speed up the process

can help reduce errors

Automation also allows you to...

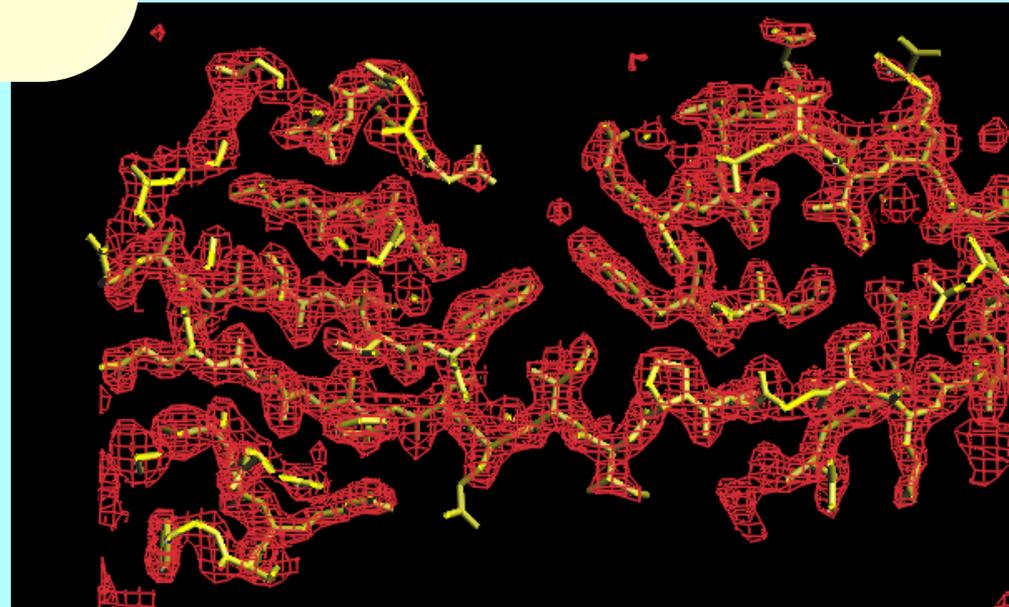
try more possibilities

estimate uncertainties



Requirements for automation of structure determination of macromolecules by X-ray crystallography

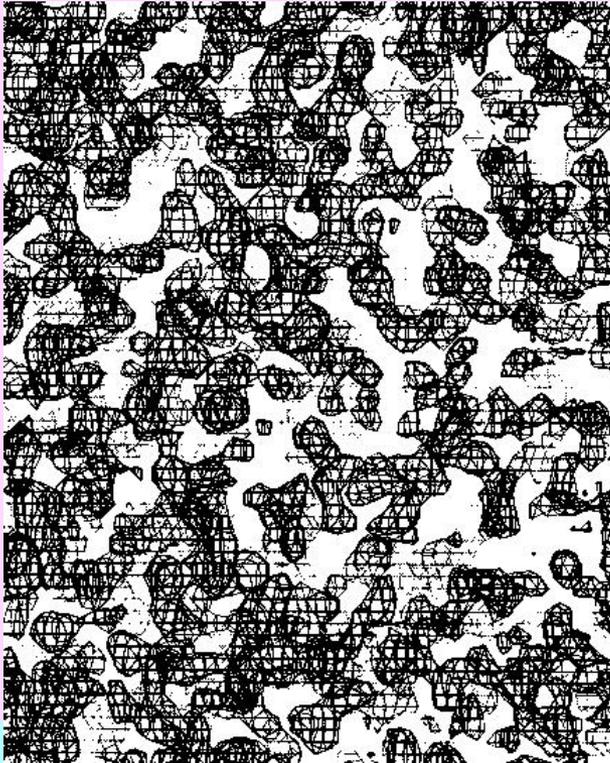
- (1) **Software carrying out individual steps**
- (2) **Seamless connections between steps**
- (3) **A way to decide what is good**
- (4) **Strategies for structure determination and decision-making**



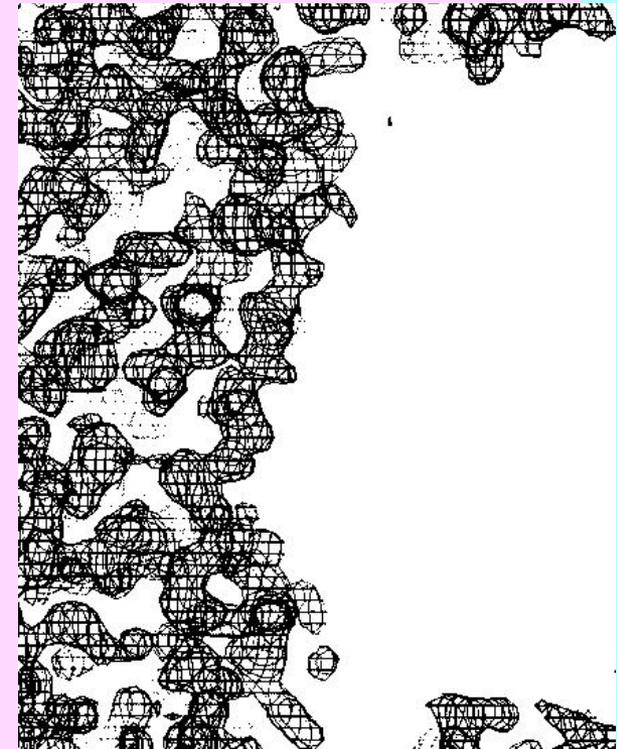
Why we need good measures of the quality of an electron-density map:

Which solution is best?

Are we on the right track?



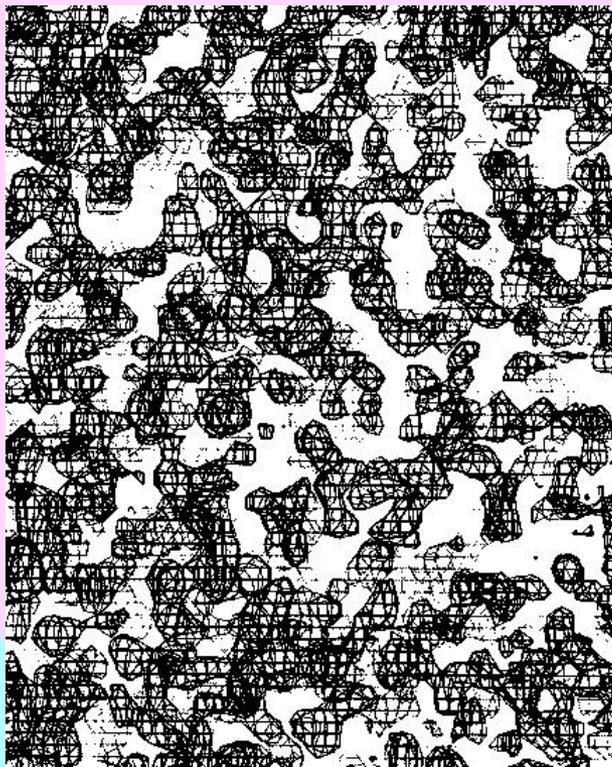
**If map is good:
It is easy**



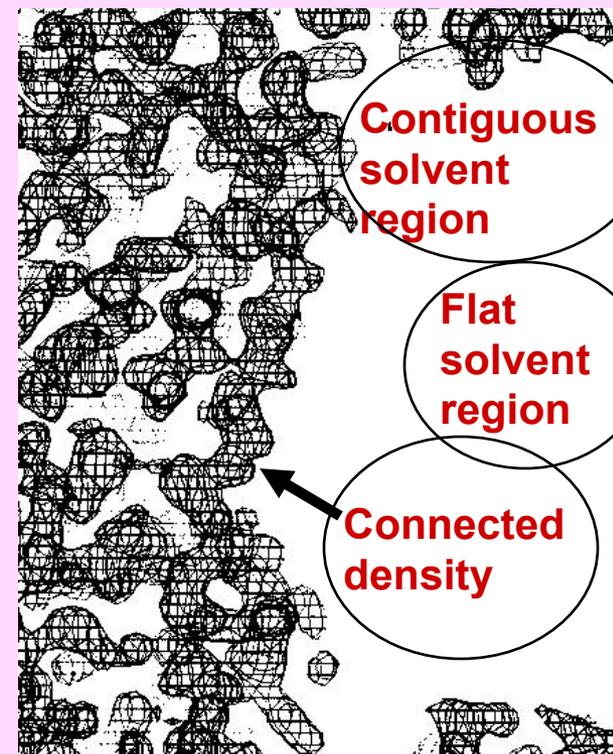
Why we need good measures of the quality of an electron-density map:

Which solution is best?

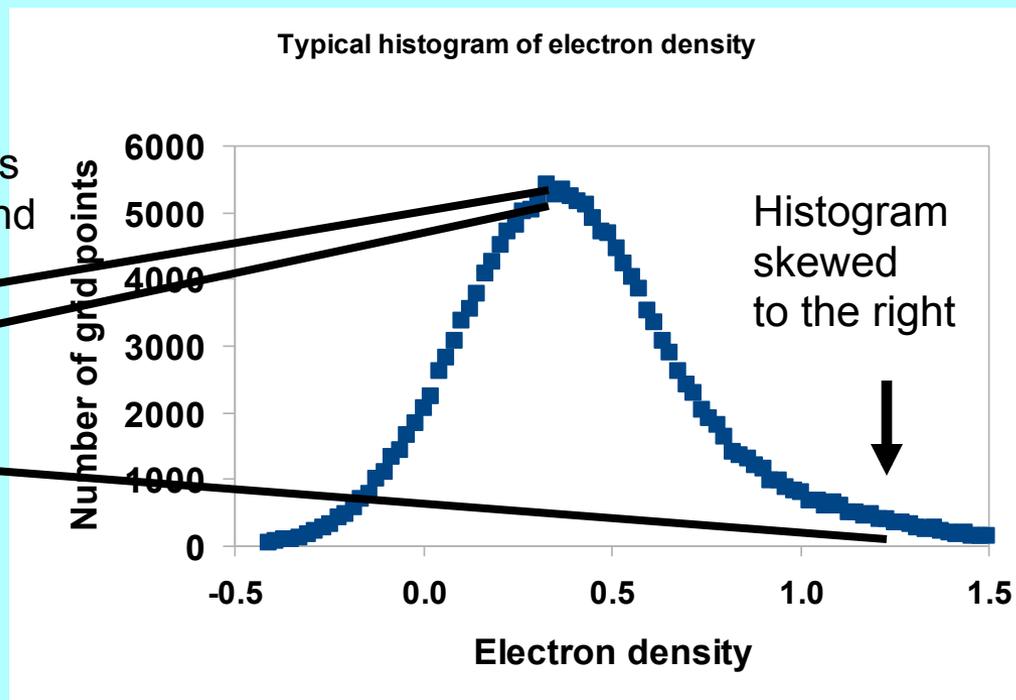
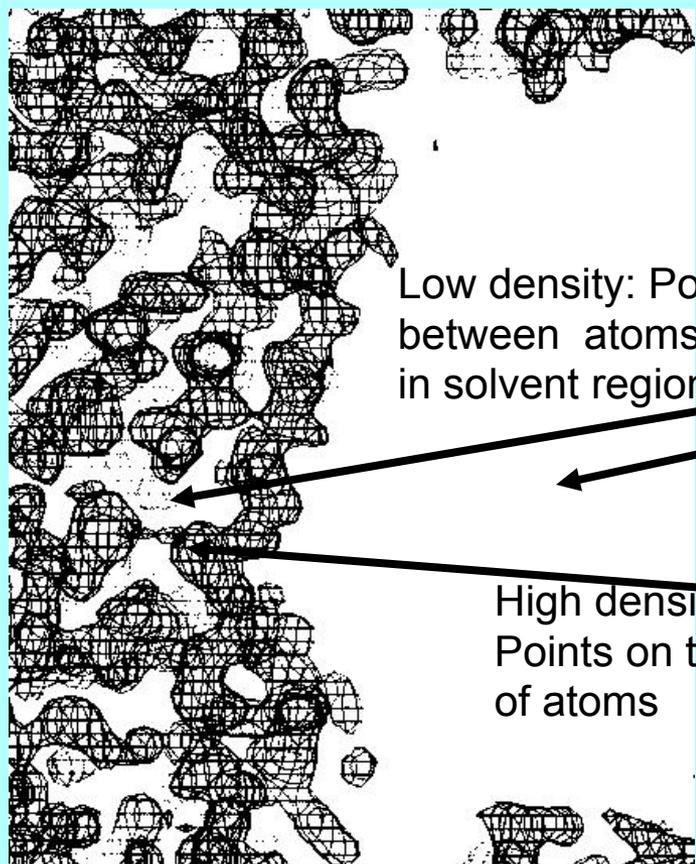
Are we on the right track?



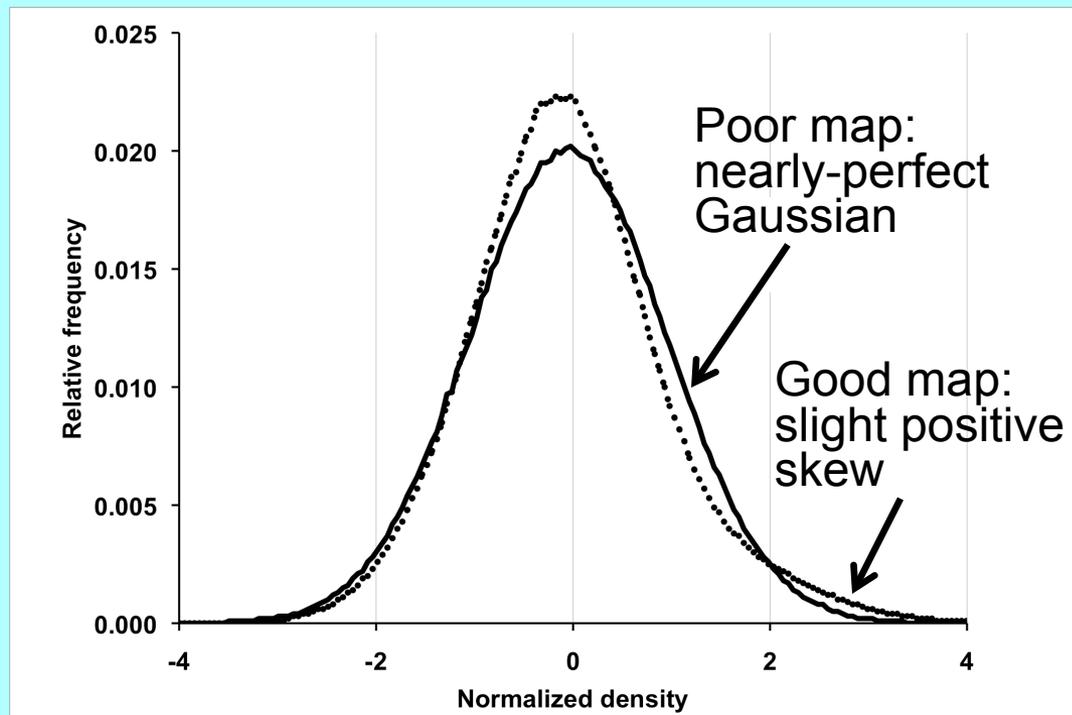
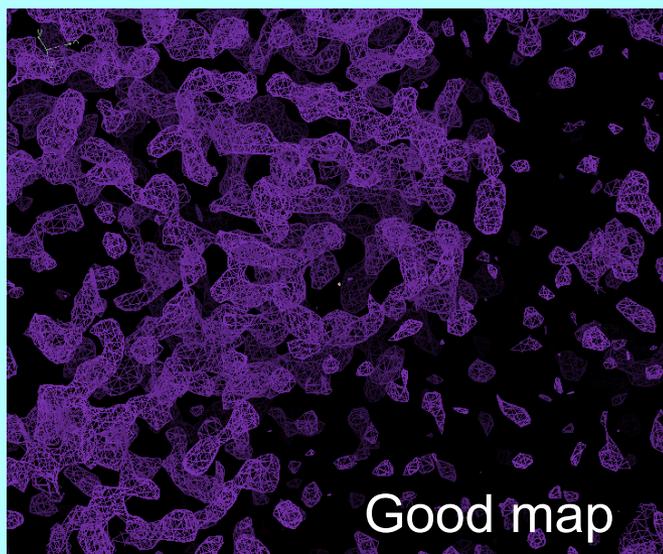
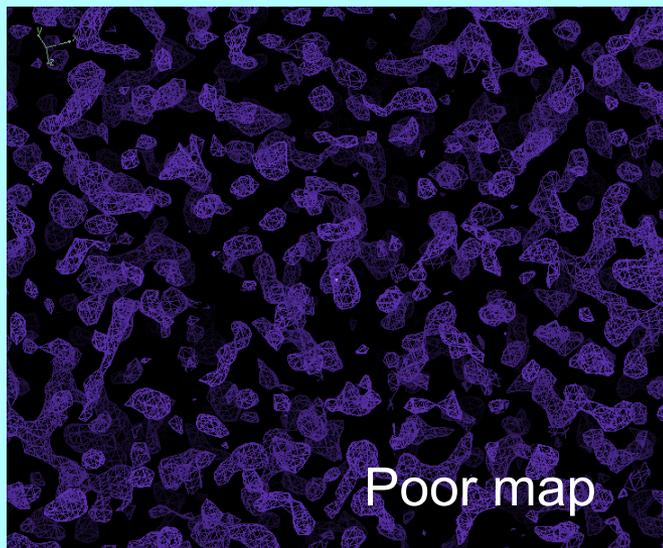
If map is good:
It is easy



Histogram of electron density values has a positive “skew”



Skew of electron density for poor and good maps



Evaluating electron density maps

<i>Basis</i>	<i>Good map</i>	<i>Random map</i>
Skew of density (Podjarny, 1977)	Highly skewed (very positive at positions of atoms, zero elsewhere)	Gaussian histogram
Connectivity of regions of high density (Baker, Krukowski, & Agard, 1993)	A few connected regions can trace entire molecule	Many very short connected regions
Correlation of local rms densities (Terwilliger, 1999)	Neighboring regions in map have similar rms densities	Map has uniform rms density
R-factor in 1 st cycle of density modification (Cowtan, 1996)	Low R-factor	High R-factor

Which scoring criteria best reflect the quality of a map?

Create real maps

Score the maps with each criteria

Compare the scores with the actual quality of the maps

Creating real maps

247 MAD, SAD, MIR datasets with final model available
(PHENIX library and JCSG publicly-available data)

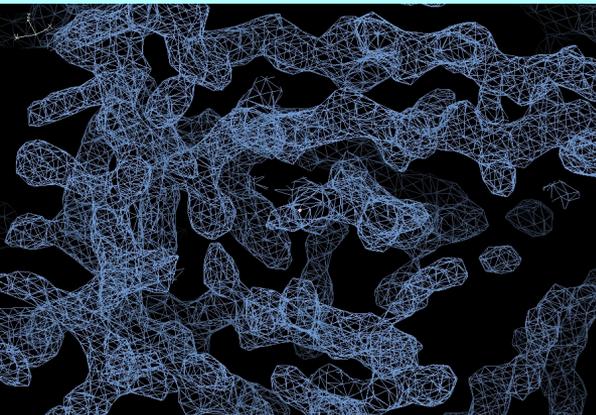
Run AutoSol Wizard on each dataset.

Calculate maps for each solution considered
(opposing hands, additional sites, including various derivatives
for MIR)

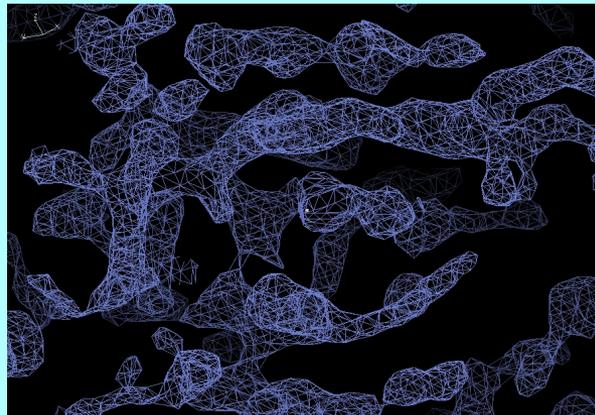
Score maps based on each criteria

Calculate map correlation coefficient (CC) to model map
(no density modification, shift origin if necessary)

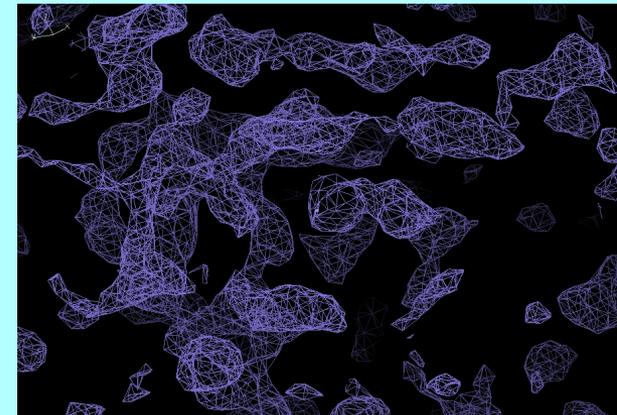
Model map
1VQB, 2.6 Å, SG C2



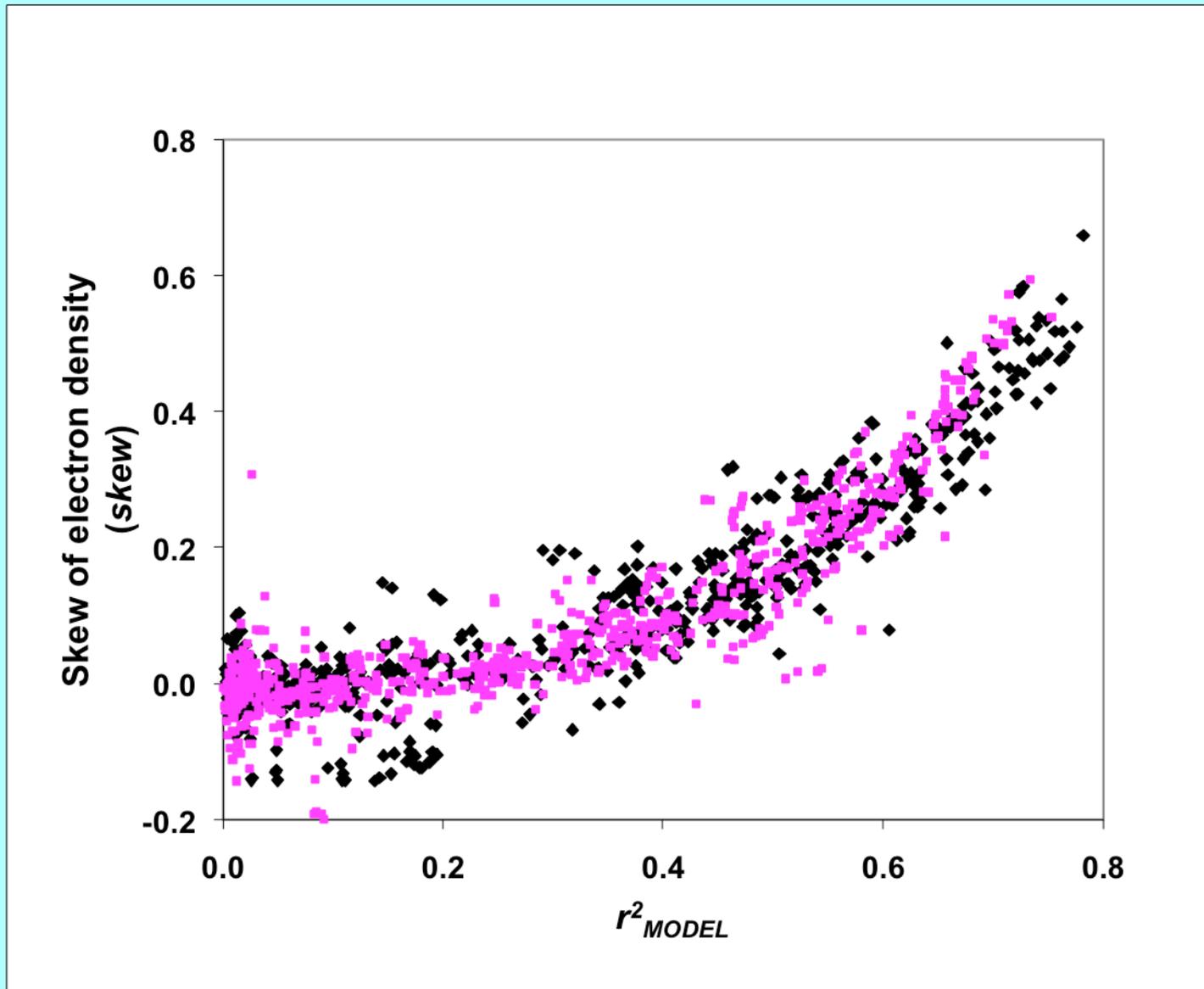
SOLVE MAD map
CC=0.62



Inverse-hand map
CC=0.55



Skew of electron density – positive skew of density values

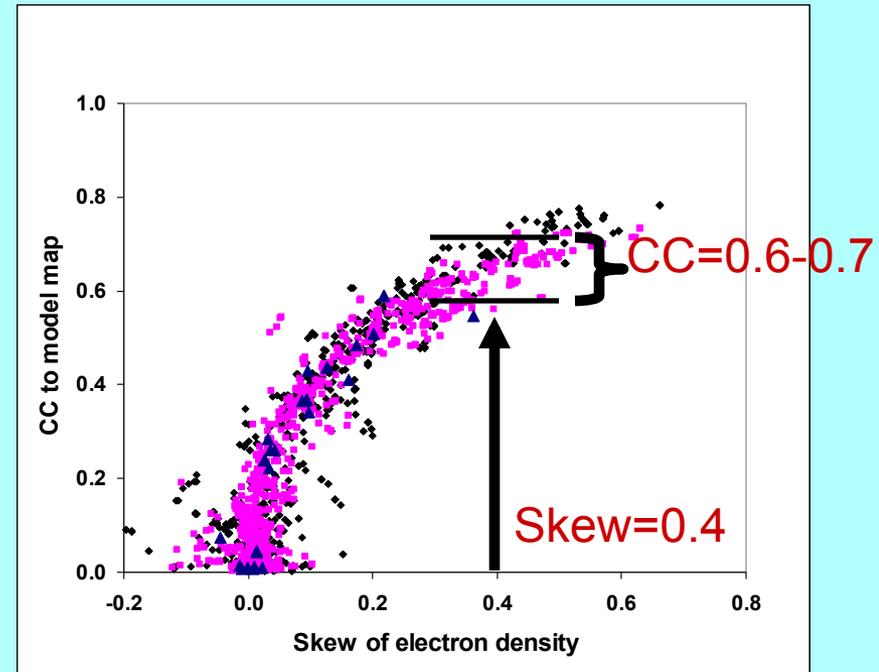
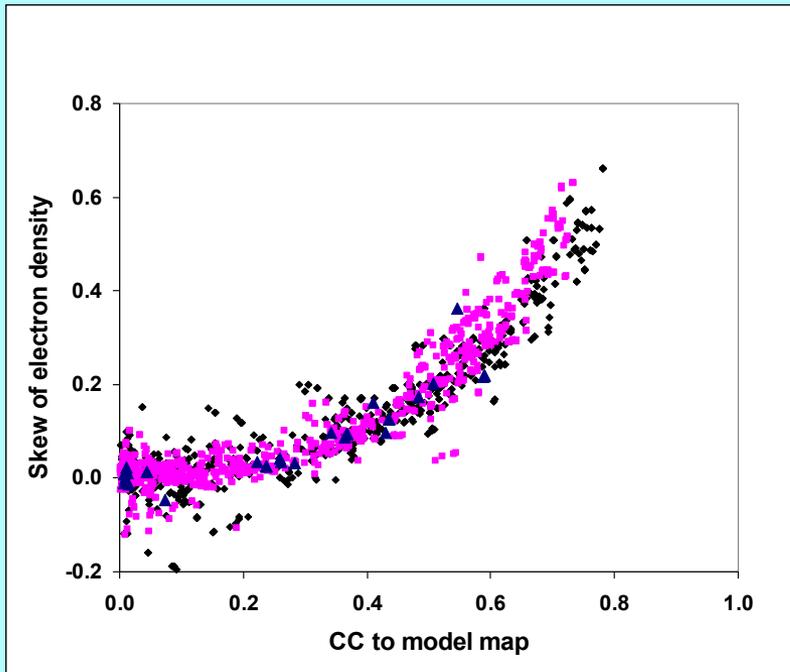


Using scoring criteria to estimate the quality of a map

Skew depends on CC

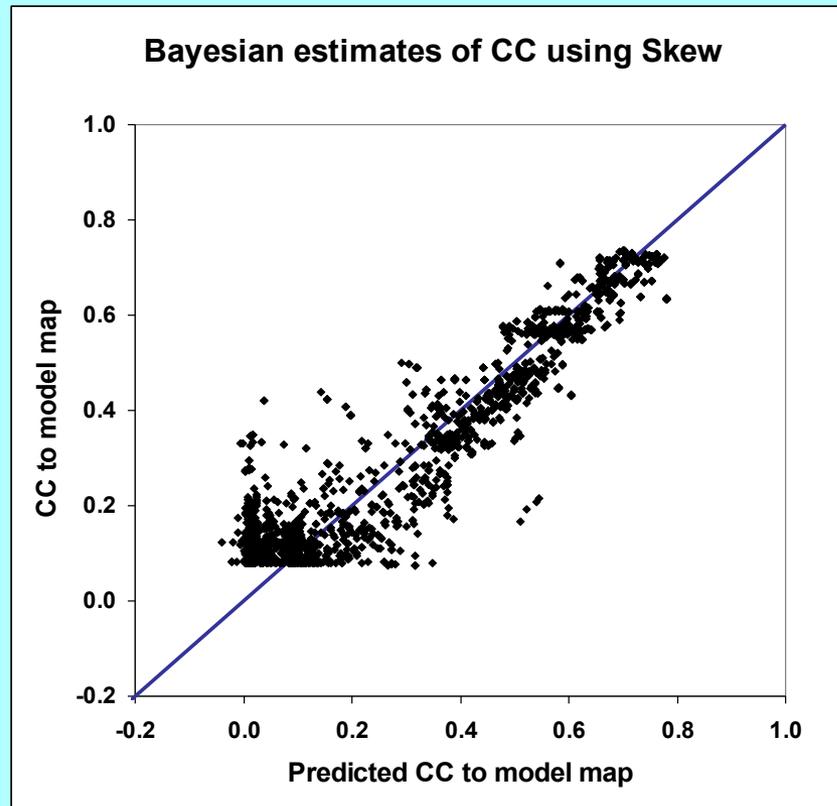


Estimate CC from skew



How accurate are estimates of map quality?

Actual
quality



Estimated quality

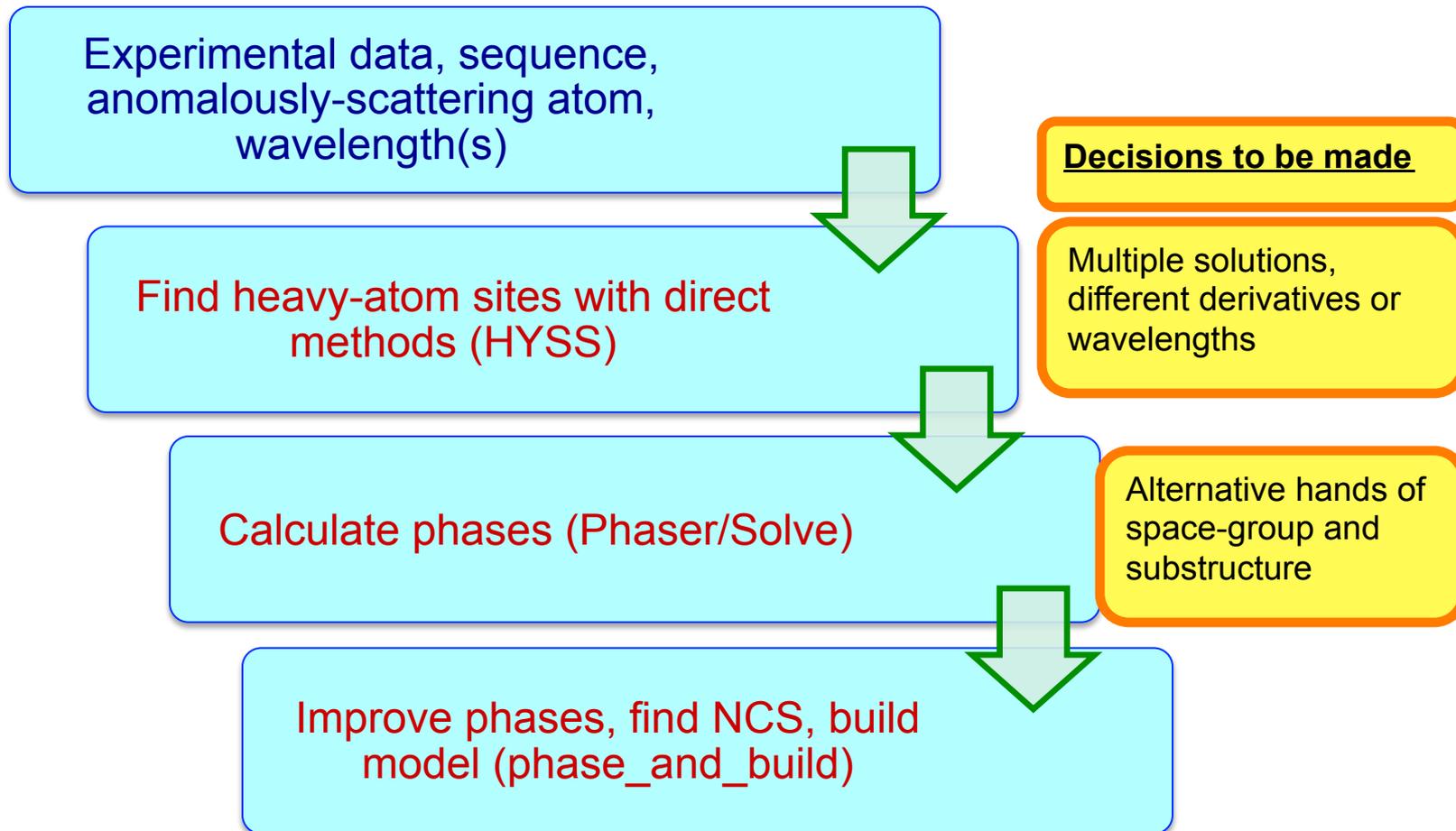
Cross-validated estimates of quality

Estimated map quality in practice

Evaluating solutions to a 2-wavelength MAD experiment
(JCSG Tm3681, 1VPM, SeMet 1.6 Å data)

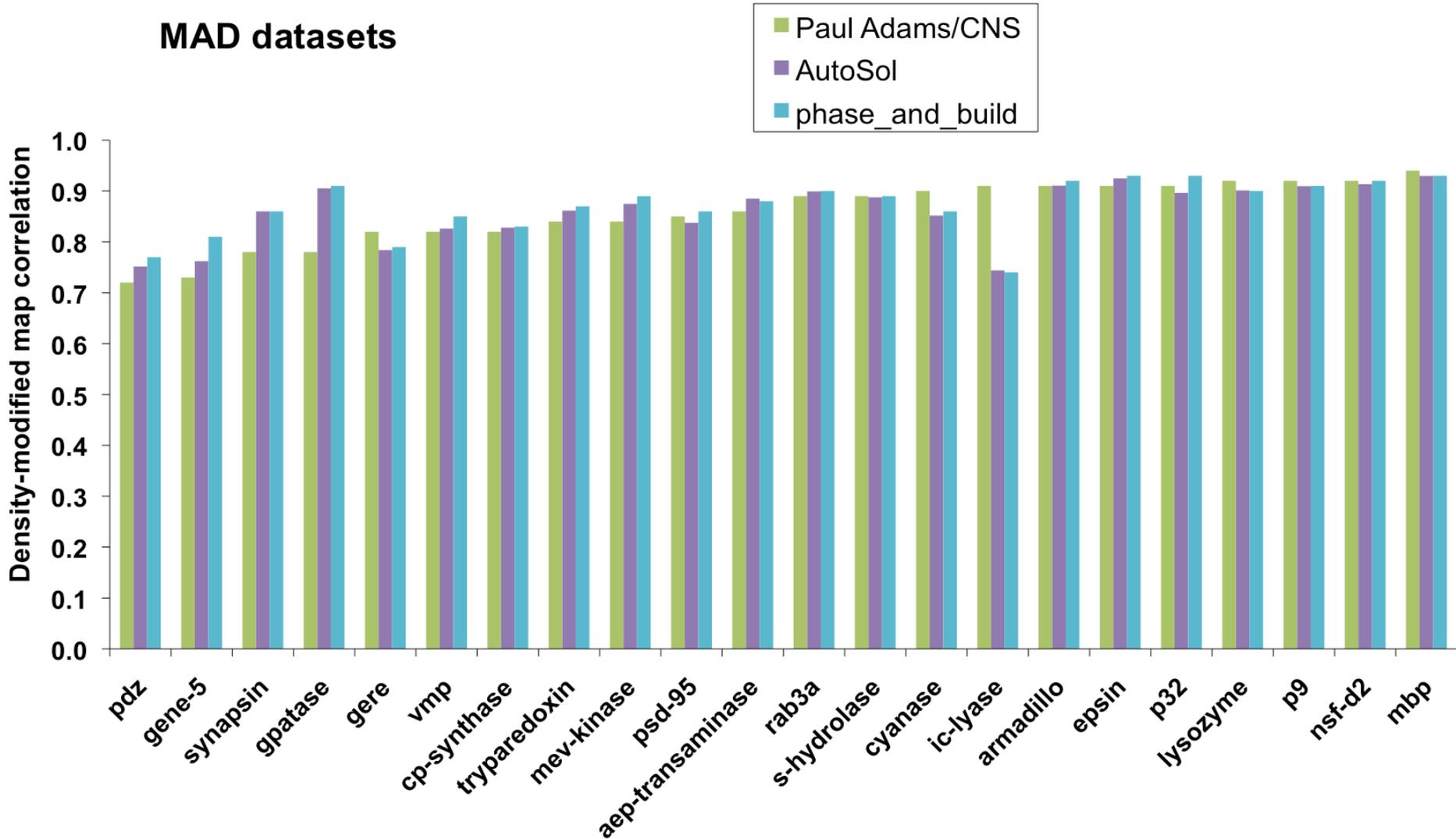
Data for HYSS	Sites	Estimated CC $\pm 2SD$	Actual CC
Peak	12	0.73 ± 0.04	0.72 ←
Peak (inverse hand)	12	0.11 ± 0.43	0.04
F_A	12	0.73 ± 0.03	0.72
F_A (inverse)	12	0.11 ± 0.42	0.04
Sites from diff Fourier	9	0.70 ± 0.17	0.69

Structure solution with phenix.autosol

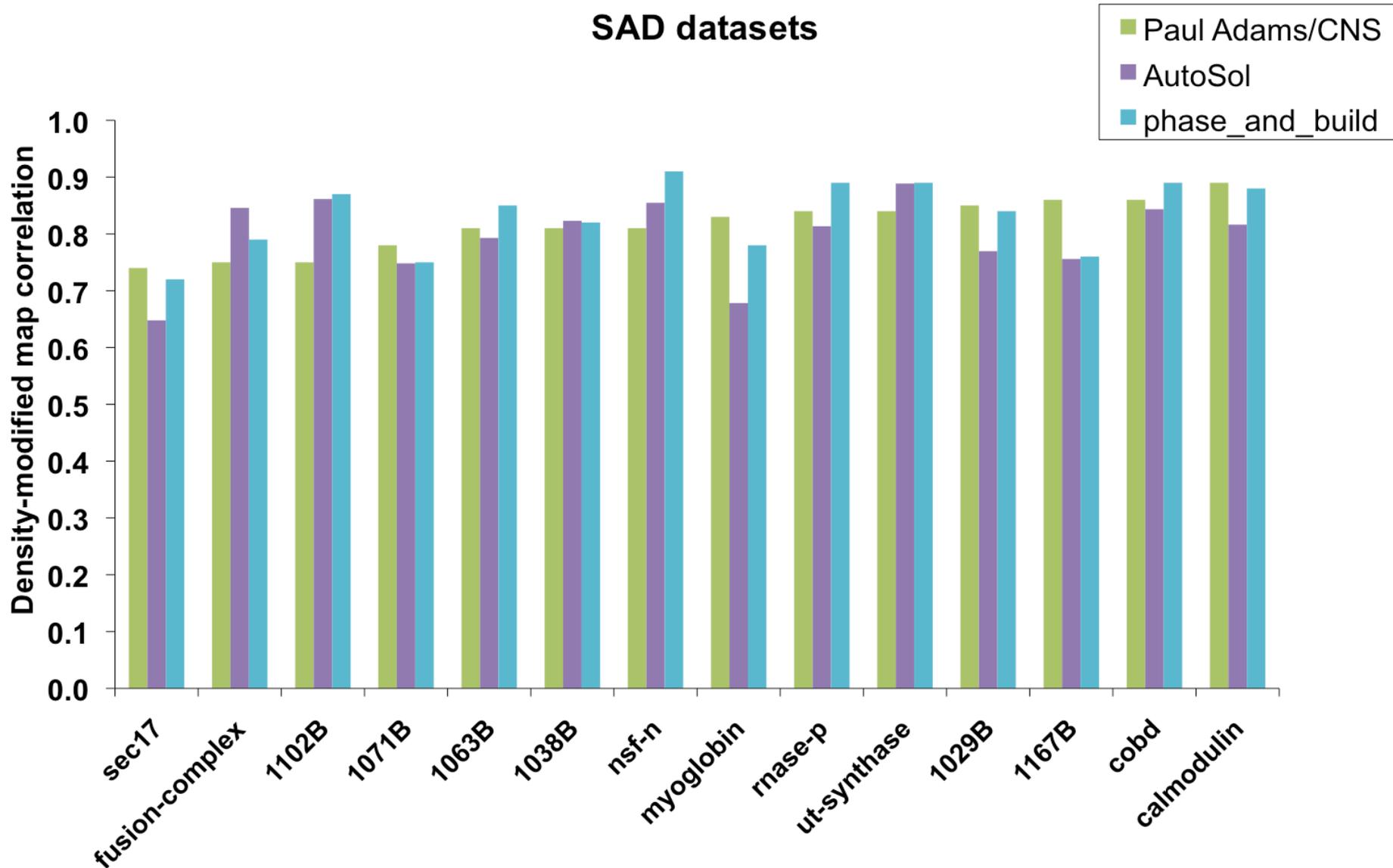


AutoSol – fully automatic tests with structure library
(MAD datasets, HYSS search, SOLVE)
RESOLVE/ phase_and_build maps

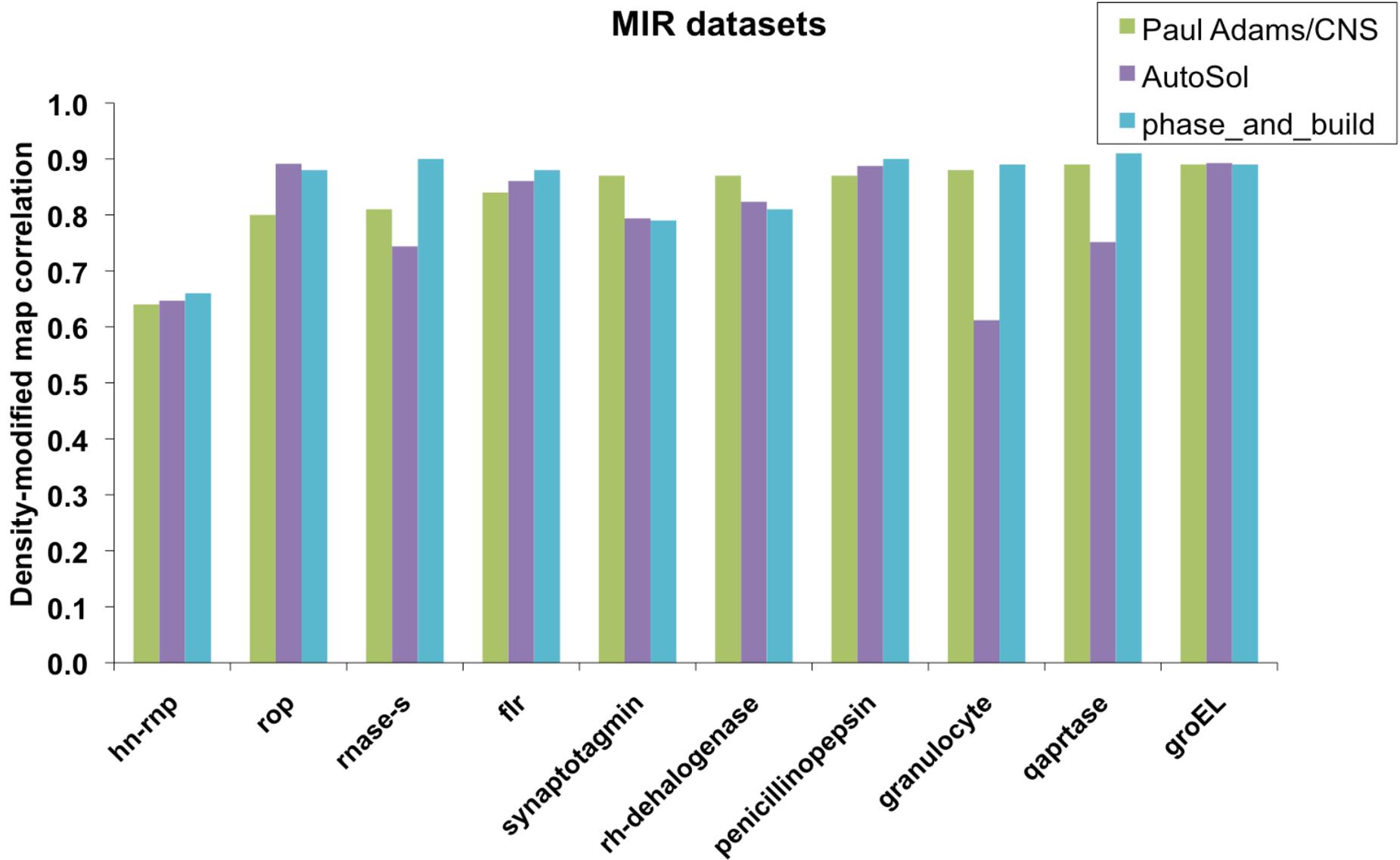
MAD datasets



AutoSol – fully automatic tests with structure library
(SAD datasets, HYSS, Phaser)
RESOLVE/ phase_and_build maps



AutoSol – fully automatic tests with structure library
(SAD datasets, HYSS, Phaser)
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The PHENIX Project



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Cambridge University



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